

ASSISTED SCRATCH REMOVALBACKGROUND OF THE INVENTION1. Field of the Invention

The present invention relates to a method for the detection and removal of line-like defects from an image. The invention relates to a process of detecting the defects using mathematical procedures and then correcting the detected defects.

2. Background of the Art

As a result of repeated handling or accident, old photographs often develop scratches. They also acquire defects resulting from material failure with age, such as cracking of surface coatings or laminations placed over the surface of the picture. The photograph may also become creased through careless handling, leading to a surface pattern resembling a scratch. Such scratch patterns are in general not simple straight lines. Instead they can be quite complex, having curved as well as straight sections with intersections and line junctions. Since the advent of inexpensive and high quality scanners, many old photographs are being digitized. Such images have high sentimental or historical value and there is a strong desire and need to digitally eliminate defects such as scratches from the images.

Much effort has been expended on restoration of cinema films that have undergone damage. An overview of the area can be found in P. Schallauer, A. Pinz and W. Haas, "Automatic restoration algorithms for 35 mm film", *Videre*, vol. 1, no.3, Summer 1999. This is an electronic journal available on-line from <http://mitpress.mit.edu/e-journals/Videre/>. Film scratches are usually substantially straight and lie along the film direction. This orientation of the scratches occurs because the scratches are usually caused by the projector transport mechanism, which lies consistently parallel to the film sprockets, and because scratches caused by trapped dirt during development can move over a few percent of the film width over several frames, again along the path of film movement during development. Compared to the arbitrary

scratches that are found in still photographic prints, the scratching of motion picture film is therefore a very restricted form of scratch. Despite this, the authors state that “from an algorithmic point of view, removing defects is an easy task in comparison to detecting them”. This position attests to two things: first, detecting even a very specific type of scratch is difficult and, second, correction of films is relatively easy because content can be taken from frames prior to and after the defect. There is, however, no such source of data for a still photo.

In digital image editing software such as, for instance, Paint Shop Pro (Jasc Software, Inc., 7905 Fuller Road, Eden Prairie, Minnesota, 55344), it is customary to provide a clone tool to repair image defects. This tool is a brush that picks up an undamaged region of the image and allows it to be painted over the damaged portion. Such a tool is effective, but requires considerable skill to produce a natural and seamless result. Moreover, the tool is not helpful when the image contains no area with undamaged content corresponding to that of the damaged area. Other software often provides a “dust and scratch removal” filter where defects are removed with a non-linear filter, such as a median filter. However, since every pixel in the image – not just scratch pixels – can potentially be modified, this technique is of very limited usefulness. For this reason, a number of proposals have been made for improved methods of scratch correction. For example, in M. Bertalmio, G. Sapiro, V. Caselles, and C. Ballester, “Image Inpainting”, *Preprint 1655*, Institute for Mathematics and its Applications, University of Minnesota, December 1999 there is described a method of filling-in scratches, which is done in such a way that isophote lines arriving at the region boundaries are completed inside. Though the method is described as fast, because it is iterative, securing the highest quality results is slow. In a paper entitled “Combining Frequency and Spatial Domain Information for Fast Interactive Image Noise Removal”, A. N. Hirani and T. Totsuka, *Proceedings of SIGGRAPH 96*, 269-276 (1996) report a method of repairing scratches in textured areas in which the user indicates a region of source texture for patching that is applied to the defect by means of frequency domain analysis. This method can be very effective in some cases but can fail in others. Neither of these improvements provide any way to

identify the scratch. In fact, both authors assume that the scratch has previously been defined somehow, presumably by hand.

However, when scratches or other defects are small and plentiful, defining the scratches prior to correcting them becomes a main objective and content of labor. For instance, the work of marking every section of every scratch is roughly equivalent to cloning or painting over the scratch in the first place, so little is gained by subsequent automation. Automatic definition of generalized scratches is not at a satisfactory level. Often knowing whether something is a scratch or not depends on understanding the content of the image, something beyond the ability of current methods dealing with arbitrary images. A need therefore exists for an assisted method of scratch removal that will allow a user to define precisely the defects in an image without needing the labor of tracing each individually and then to perform a correction of the so-defined scratches.

### **SUMMARY OF THE INVENTION**

An aspect of this invention is to provide a means of removing line like defects from an image, without a requirement for manually marking regions upon the image as regions that contain defects, comprising detecting defects with known methods of detecting lines with specific characteristics and then correcting the detected defects. It is a further aspect of the invention to provide a means of removing line like defects from an image, without a requirement for manually marking regions upon the image as regions that contain defects, comprising detecting the defects using a local radial angular transform and then correcting the detected defects.

### **BRIEF DESCRIPTION OF THE FIGURES**

Figure 1 shows one distribution of groups of pixels in a hexon.

Figure 2 shows another distribution of groups of pixels in a hexon.

Figure 3 shows yet another distribution of larger groups of pixels in a hexon.

## **DETAILED DESCRIPTION OF THE INVENTION**

The term line defect or line-like defect is a meaningful term within the practice of the present invention. There are many different types of defects, ranging from dropouts (single pixel defects or failures), color shifts, fading, actual damage to an original figure, etc. No single process sequence can address all of the forms of defects or damage, so individual systems, software, and processes must be devised to address the different defects. Line or line-type defects constitute defects or image damage that is defined a series of attached or adjacent points that extends over a distance comprising at least a multiple number of pixels (e.g., at least 2, at least 4, at least 6, at least 10, at least 14 pixels, and defining even macroscopic dimensions and visible distances (e.g., at least 1 mm, at least 2 mm, at least 3 mm, at least 5 mm and more) in which there is damage in the image or defects in the data of the image. The 'line' or 'line-type' defect does not have to be a perfectly straight line, but may be jagged, curvilinear, discontinuous or the like. A line may be generally assumed to have a dimension of width that is small (less than 5%, less than 3%, and even less than 1%) of the largest dimension of the image.

A digital image comprises a collection of picture elements or pixels arranged on a regular grid. A gray scale image is represented by a channel of specific brightness values at individual pixel locations. Such a channel may also be represented as a color palette, for example a palette containing 256 shades of gray. A color image contains several channels, usual three or four, that are used to describe the color at a pixel. For example, there may be red, green and blue (RGB) channels, or cyan, magenta, yellow and black (CMYK) channels. Each channel again contains brightness values representing the color at each pixel. Equally, the image may be represented in a color space having a lightness channel along with other channels directly or indirectly representing the hue and saturation components of color. Non-limiting examples of formats for color space definition include the HLS, HSV, YIQ, YUV, YES, CIE L\*u\*v\* and CIE L\*a\*b\* color spaces. All the aforementioned channels are suitable for the practice of the invention.

Current methods of removing from a single still image line-like defects (such as for example, scratches) all have one thing in common, namely that the location of the defect must be marked upon the image in some way. The marking may take several forms. For instance, the pixels forming the defect may be marked individually, for example by means of conventional selection tools such as freehand or lasso tools that outline a selection area or by means of a magic wand tool that selects pixels by color similarity to a target pixel. Subsequently, selected pixels are modified to reduce the visibility of the defect. Alternatively, the approximate region of the defect may be designated, for example by placing a virtual frame upon the image containing at least the pixels of the defect. Subsequently, in the region of the image restricted to the contents of the virtual frame, an attempt is made to classify pixels into defect and non-defect pixels by some means, whereupon the defect pixels are modified to reduce the visibility of the defect. Yet another approach marks the defect indirectly, in the sense that an operator paints a correction directly over the defect pixels but substantially not over other, undamaged pixels of the image. In this case, image pixels are modified unconditionally without regard to their being defect or non-defect pixels and restriction of correction to only defect pixels is accomplished by the operator controlling the location, size and other characteristics of the conceptual brush with which painting is accomplished. With a single defect isolated from other similar image features or a small number of such defects these known approaches to correction of line-like defects may be adequate. However, with many small defects in different regions of the image such methods are impractically laborious, especially when it is recognized that digital images commonly contain at least thousands of pixels and often many millions of pixels. Currently this problem is without a solution.

It is known to clone over scratches (line defects) or, once they are manually marked on the image, to repair them with in-painting, etc. It is also known to manually define the approximate spatial location of a scratch. These approaches require the scratch to be painted on or defined in the spatial domain and not according to its properties. It is known that various software asserts an ability for “dust and scratch” removal, though in practice it cannot cope with extended lines (i.e., long lines). This “dust and scratch” removal is accomplished, for instance, by processing an image with rank order filters or filters with internal thresholds (e.g., despeckle filters), which are applied to every pixel in the image without regard to whether it is a defect or non-defect pixel and so can damage the image features that should remain unmodified. It is known to automatically correct lines in movie frames according to their properties in dependence on information in prior and subsequent frames such as, for example, the absence of a scratch in a subsequent frame or the known location of a scratch in a previous frame. There is no known method of scratch removal in single still images that permits the scratch to be defined and then corrected, where the definition rests on properties other than the spatial location of the scratch.

A novel aspect of this invention is the conception of and provision of means for correcting line-like defects in a single still image that does not require the location of the defect pixels to be manually marked on the surface or area of the image to be corrected. Any conventional means of detecting lines may be used for identifying line-like defects provided they provide a means of selectively detecting lines with different characteristics. Such characteristics may, for example, include the contrast of the line with respect to its surroundings, the orientation of the line, the sharpness of the line edges, the width of the line and the like. Methods of line detection suitable for the practice of this invention may be found, for example, in C. Steger, “An Unbiased Detector of Curvilinear Structures”, *Technical Report FGBV-96-03*, Forschungsgruppe Bildverstehen, Informatik IX, Technische Universität München, Germany, 1996 and references therein. Schemes for line detection range from a concentration on local brightness differences in the image, through detection of lines as objects having parallel edges, to more sophisticated techniques. Some of these more sophisticated techniques use the curvature of the

brightness of the image for estimation of lines using contours, ridges and ravines and locally fitting the curvature of the image. These methods may be used for the practice of the invention but many of them are of considerable computational complexity. Another aspect of the invention is, therefore, to provide an additional and exemplary method of line defect detection that may be used in a process of correcting line-like defects in an image by detecting these defects without a requirement for manually marking regions upon the image as regions that contain defects, and then correcting the detected defects. This exemplary method is rapid and is capable of characterizing lines by means that include characterization according to whether they are dark or light, by contrast with respect to their surrounds, by width, by length, by orientation and by sharpness of edge definition. The method is based on the local radial angular transform. The listed characteristics are valuable for differentiating a line-like defect, such as scratch, from other image features and this method is accordingly a preferred method in the practice of the invention.

A co-pending U.S. Patent application and bearing attorney's docket number 1202.018US1 and titled LOCAL RADIAL ANGULAR TRANSFORM PROCESS discloses a local radial angular transform utilizing a hexagonal structure, termed a hexon, that is overlaid over the pixels of the image. Procedures for performing this overlay are described in detail in the co-pending application, which is incorporated herein in its entirety. The hexon consists of a central group of pixels surrounded by six groups of pixels arranged in approximate hexagonal symmetry about the central group. Figure 1 of this application shows two orientations of one such hexagonal arrangement of pixel groups. Figure 2 shows a different arrangement of pixel groups in which the groups do not touch. Figure 3 shows a similar arrangement to that of Figure 2 but with larger pixel groups.

The numbers shown in the figure identify each of the pixel groups and appear as subscript labels in the following discussion. In an image having any particular brightness values at its pixels, the mean brightness values of the six groups surrounding the central group may be represented as a vector  $B = (B_1, B_2, B_3, B_4, B_5, B_6)^T$ , where the superscript

T denotes a transpose operation that converts a row matrix to a column matrix. A Local Radial Angular (LORA) transform  $L_c$  is defined as  $c = RB$ , where  $c = (c_1, c_2, c_3, c_4, c_5, c_6)^T$  is a vector of transformation coefficients. R is a six by six square matrix whose elements are formed according to:

$$R_{km} = (1/6) \exp[i (k - 1) (m - 1) \pi/3] \quad (k, m = 1, 2 \dots 6)$$

where  $i$  is the imaginary unity (i.e., the square root of  $-1$ ),  $\pi$  is the ratio of the circumference to the diameter of a circle, and  $k$  and  $m$  are the row and column indices of the matrix elements. The explicit form of  $c_3$  is given by:

$$c_3 = (0.5/6) (2B_1 - B_2 - B_3 + 2B_4 - B_5 - B_6) + i (0.5/2) (B_2 - B_3 + B_5 - B_6)$$

so that the real and imaginary components of  $c_3$  are given, respectively, by:

$$\begin{aligned} \text{Real}(c_3) &= (0.5/6) (2B_1 - B_2 - B_3 + 2B_4 - B_5 - B_6) \\ \text{Imaginary}(c_3) &= (0.5/2) (B_2 - B_3 + B_5 - B_6) \end{aligned}$$

The magnitude of the modulus  $|c_3|$  of the transformation coefficient  $c_3$  has been found to be an indicator of the presence of a line-like feature in the image lying under the hexon superimposed over the image. There are two values of  $|c_3|$  corresponding to the orientations 1 and 2 shown in Figure 1. These separate values  $|c_{13}|$  and  $|c_{23}|$  may be combined into a single value  $|c_3|$ , for instance by taking the larger of the two, taking the square root of the sum of the squares, weighted averages, and the like. The magnitude of  $|c_3|$  can be considered the strength of a line response in the image. It is also possible to define a different hexon response,  $\delta_3$ , which is a measure of the line purity. This hexon response quantity is defined as:

$$\delta_3 = 2 |c_3|^2 / \sum_{k=2}^{k=6} |c_k|^2$$



Other definitions involving weighted functions of the coefficients  $c_k$  are also possible. Since coefficients other than  $c_3$  respond to image features that are not lines,  $\delta_3$  is a measure of the degree to which the  $c_3$  response represents a line. There are separate values,  $\delta_{13}$  and  $\delta_{23}$ , of this measure for hexon orientation 1 and hexon orientation 2, respectively (see Figure 1). These may be combined into a single value of  $\delta_3$  by any convenient means, for instance by using the larger value. The  $c_3$  coefficient responds to lines that are both dark and light with respect to the background upon which they lie and by default both types of lines are detected. However, it is also possible to selectively detect only light lines or only dark lines. This may be achieved in various ways. For example, the mean brightness or channel value at the quasi-pixels lying closest to the line may be compared to the value of  $c_1/6$ . Alternatively, the lightness or darkness of a line may be estimated from the real and imaginary parts of the  $c_3$  coefficient by comparison to thresholds  $T_1$  and  $T_2$  according to the following logic:

if  $|\text{Imaginary}(c_3) / \text{Real}(c_3)| \leq T_1$  and  $\text{Real}(c_3) > T_2$  then Light  
 if  $|\text{Imaginary}(c_3) / \text{Real}(c_3)| > T_1$  and  $\text{Real}(c_3) < T_2$  then Light  
 if  $|\text{Imaginary}(c_3) / \text{Real}(c_3)| \leq T_1$  and  $\text{Real}(c_3) < T_2$  then Dark  
 if  $|\text{Imaginary}(c_3) / \text{Real}(c_3)| > T_1$  and  $\text{Real}(c_3) > T_2$  then Dark

While the value of  $T_1$  depends on the detailed geometry of the hexon, a preferred value of the threshold  $T_1$  is from greater than about 0 to less than about 0.57. An especially preferred value is about 0.07 to about 0.41, with a most especially preferred value of about 0.3. The preferred value of  $T_2$  is about 0. In this way either light or dark lines may be separately detected.

Referring to Figure 1, for orientation 1 of the hexon, the orientation angle  $\theta_1$  of the line-like feature measured anticlockwise from the image horizontal is:

$$\theta_1 = \arctan[\text{Imaginary}(c_3) / \text{Real}(c_3)]$$

while for orientation 2 of the hexon, the orientation angle  $\theta_2$  of the line-like feature is:

$$\theta_2 = \arctan[\text{Imaginary}(c_3) / \text{Real}(c_3)] + \pi/2$$

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T09020.070601

The angle  $\theta$  is a continuous function of  $c_3$ , giving rise to angles in the interval 0 to 180 degrees, but for exact hexagonal symmetry, the highest accuracy in  $\theta$  is obtained near (but not necessarily exactly at) angles of 0, 30, 60, 90, 120 and 150 degrees. In practice, the  $\theta$  angles of highest accuracy deviate slightly from these values because the requirement to map the hexon to pixels on a square grid leads to slight distortions of the hexagonal symmetry. For example, in the arrangement shown in Figure 1, the angles of highest accuracy in  $\theta$  are 0, 26.6, 63.4, 90, 116.6 and 153.4 degrees. The estimate of the orientation angle of the line-like feature can also, for instance, be improved by computing the angle as an average of  $\theta_1$  and  $\theta_2$  weighted by the  $|c_3|$  responses of the two orientations of the hexon.

15 Additionally, when a straight or curved line segment of specified width has been detected it is possible to count the number of pixels in the line segment. By division of the pixel count by the line width it is possible to approximately estimate the length of the line segment. Accordingly, line segments can be selected according to length, for example by requiring the length to fall between two thresholds. Further, by means of suitable choices of hexon as disclosed in co-pending U.S. Patent application bearing attorney's docket number 1202.018US1 and titled LOCAL RADIAL ANGULAR TRANSFORM

20 PROCESS, for example a hexon of non-contiguous groups as shown in Figure 2, it is possible to detect lines with periodic spacing such as dashed lines.

25 Thus, overall, there are at least seven means of line analysis –  $|c_3|$ , the line strength metric,  $\delta_3$ , the line purity metric,  $T_1$  and  $T_2$ , the thresholds determining lightness or darkness of the line, the line orientation, line length, and the size and shape of the pixel group used to form the hexon, which determine the width and continuity of the line to be detected. The size of the pixel group and the size of the associated hexon determine the scale or width of the line that will be detected. The strength metric responds, for instance, to the contrast of the line relative to its surroundings. High contrast leads to a high value of  $|c_3|$  and low contrast to a low value of  $|c_3|$ . The purity metric responds, for instance, to how well defined the line is. Lines with sharp edges yield high values of  $\delta_3$ , while lines

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with blurred, ill-defined edges give smaller values of  $\delta_3$ . It has been found that the  $|c_3|$  and  $\delta_3$  metrics, along with the  $T_1$  and  $T_2$  thresholds and the hexon size, can be used in combination to effectively delineate scratches in an image prior to correction. This is because by means of these criteria it is possible to select from the image only line-like structures of a specified lightness relative to their surroundings in conjunction with a specified contrast, line definition and line width.

One of the benefits of the invention, is that it can be used to isolate scratches (etc.) in an image for correction without the labor of having to point out and trace each individual scratch. To do this successfully, one has to differentiate line-like scratches from authentic line-like image details. An embodiment based on the local radial angular transform includes differentiation in these ways:

1. Whether the line is light or dark with respect to its surroundings.
2. How well defined the line is. This is the dependence on  $\delta_3$ , which generally measures the degree to which the detected object is like a line and not like something else (e.g. a semi-plane). It may be thought of as roughly determining how fuzzy the edges of a line have to be before something is not considered a line.
3. Determining what is the contrast of the line relative to its surroundings. This is the dependence on  $|c_3|$ , which may be specified for instance as a range between two thresholds.

However, the invention may actually be practiced without reference to the preferred LORA functions when an alternative method of line detection is used.

The method can be practiced as removing line defects from a still image by providing image data in digital form, detecting line defects in the image within a specified range of widths without manually designating the spatial location of the line defects, and adjusting the image data to correct the detected line. It can also be described as a method for removing line defects from a still image by providing image data in digital form, detecting line defects in the image of a specified brightness, either higher or lower, compared to their surrounding without manually designating the spatial location of

the line defects, and adjusting the image data to correct the detected line. It may alternatively be described as a method for removing line defects from a still image by providing image data in digital form, detecting line defects in the image of a specified contrast compared to their surroundings without manually designating the spatial location of the line defects, and adjusting the image data to correct the detected line or as a method for removing line defects from a still image by providing image data in digital form, detecting line defects in the image of a specified range of sharpness without manually designating the spatial location of the line defects, and adjusting the image data to correct the detected line.

The invention envisages broad scope for use of the aforementioned quantities for definition of line-like defects such as scratches, cracks or the like. For example,  $|c_3|$  or  $\delta_3$  or both may be subjected to an upper threshold that must be exceeded at a pixel for it to be marked as a defect requiring correction. Equally,  $|c_3|$  or  $\delta_3$  or both may be subjected to a lower threshold that cannot be exceeded at a pixel for it to be marked as a defect requiring correction. Additionally  $|c_3|$  or  $\delta_3$  or both may be required to have values at a pixel lying between two thresholds or outside both thresholds to qualify the pixel as a defect candidate. Alternatively, values of  $|c_3|$  or  $\delta_3$  or their ranges may be taken from pre-existing tables, for instance representing specific types of commonly occurring defects. Further suitable values or ranges of  $|c_3|$  or  $\delta_3$  for defining a defect may be selected on the basis of the statistical analysis of the distribution  $|c_3|$  or orientation of  $c_3$  or the distribution of  $\delta_3$  or any, all or several of the aforementioned within the entire image or some region of interest within the image. Defect pixels can also be selected on the basis of whether they are dark or light compared to their surroundings or both. After defect pixels are identified using the LORA method it is also possible to perform supplementary searches, for example using a conventional search for line sections based on identifying three similar contiguous pixels within a 3 by 3 pixel window. Such or similar searches can be helpful in precisely delineating the ends of a scratch when the hexon is large.

The pixels that are members of a defect may be marked by any convenient means known in the art. For example, they may be represented as a list or a mask or by means of flags

or by line segment encoding or by run length encoding or by a chain code or by other means. The defect may also be represented by a mathematical function fitted to the location of some or all of the pixels marked as part of the defect.

- 5 The size of the pixels groups making up the hexon and the size and form of the hexon itself may be chosen over wide limits as described in the above-identified co-pending U.S. Patent Application bearing attorney's docket number 1202.018US1. The image may be analyzed with one or more kinds of hexon with identical or different choices for the aforementioned thresholds. Output of the different hexons may be analyzed
- 10 independently or jointly, optionally in a hierarchy with respect to scale.

After they have been defined, the defect pixels may be corrected by any method known in the art. For example, the pixel may be replaced by the average or weighted average of pixels in its neighborhood, preferably excluding other defect pixels. The output of a top

15 hat or rolling ball filter may also be used. Non-linear filters such as the median filter or other rank leveling filters may be employed. Adaptive filters are another alternative, such as the double window modified trimmed mean filter described in "Computer Imaging Recipes in C", H. R. Myler and A. R. Weeks, Prentice-Hall, 1993, p.186ff. The defect may also be corrected by the use of morphological operations such as erosion or dilation,

20 selected on the basis of the lightness or darkness of the defect relative to its surroundings. Combinations of these operations in the form of morphological opening and closing are also possible. The defect may also be removed by interpolation such as by linear or quadratic interpolation. Other interpolation methods, for example such as the trigonometric polynomial technique described on-line by W. T. Strohmer in "A

25 Levinson-Galerkin algorithm for trigonometric approximation" at <http://tyche.mat.univie.ac.at/papers/inpress/trigappr.html> or the multivariate radial basis technique described on-line by H. Zatschler in "M4R Project – Radial Basis Functions" at <http://www.doc.ic.ac.uk/~hz3/m4rproject/m4rproject.html> may also be used.

Interpolation may also be accomplished by fitting a surface such as a plane or a parabola

30 to the local intensity surface of the image. In color or multichannel images, information from a defective channel may be reconstructed using information from the remaining

undamaged channels. The defect may also be repaired using the method of Hirani as described in A. N. Hirani and T. Totsuka, *Proceedings of SIGGRAPH 96*, 269-276 (1996). Alternatively the repair may be effected by ‘inpainting’ as discussed in M. Bertalmio, G. Sapiro, V. Caselles, and C. Ballester, “Image Inpainting”, *Preprint 1655*,  
 5 Institute for Mathematics and its Applications, University of Minnesota, December 1999 or by the more recent variational method described in C. Ballester, V. Caselles, J. Verdera, M. Bertalmio and G. Sapiro, “A Variational Model for Filling-In” available on-line at <http://www.ceremade.dauphine.fr/reseaux/TMR-viscosite/preprints.html>. Additional techniques are described in T. F. Chan and J. Shen, “Morphology Invariant  
 10 PDE Inpaintings”, *Computational and Applied Mathematics Report 01-15*, UCLA, May 2001 and T. F. Chan and J. Shen, “Non-Texture Inpainting by Curvature-Driven Diffusions (CDD)”, *Computational and Applied Mathematics Report 00-35*, UCLA, September 2000. After repair, noise may optionally be added to the corrected area to further disguise the correction. The amount of such noise may be predetermined or, for  
 15 instance, computed from the local or global properties of the image.

While the invention will be described with respect to a specific embodiment, it is understood from the foregoing that other embodiments are possible and will be apparent to those skilled in the art. For clarity the procedure will be described as a series of steps  
 20 but it will be understood that the order of the steps can be adapted to the needs of the application. For instance, certain steps may be combined or separated into sub-steps.

The following example implementation is designed to remove one- or two-pixel wide scratches, a size chosen as the largest that can be corrected in small images, such as those  
 25 used the World Wide Web, without disrupting image detail. The shape, size and conformation of the hexon detector that is used is shown in Figure 1. The top left pixel in the group marked 0 is positioned over the test pixel. The procedure comprises a defect detection phase and a defect correction phase.

### Step 1 – Initialization

The operator is given a means of specifying whether to remove dark or light scratches or both, for example by means of checkboxes. In the case of a color image the color channels, such as RGB, are converted to a color space in which there is an axis representative of lightness. Suitable color spaces include YIQ, YUV, YES, CIE  $L^*u^*v^*$ , CIE  $L^*a^*b^*$ , HLS, HSV. For a gray scale image the terms ‘dark’ and ‘light’ are unambiguous. For a color image, however, these terms only have meaning with respect to variation along the lightness axis. Accordingly, the search for the scratch is conducted only in the lightness channel, i.e., in Y, L or V for a color image, even though all channels are corrected. Further, the operator specifies the  $\delta_3$  threshold, for instance with a slider or a numeric edit control. A preferred range of this threshold is from about 0.3 to 0.99 with a most preferred range from about 0.5 to 0.9. Within the preferred range of the threshold small differences are not critical. Thus, in practice, the range may be represented by a limited number of thresholds lying in this range. For example, the operator may be allowed to choose from  $\delta_3$  thresholds of about 0.7, 0.8 and 0.9. These values may be given numerically, or as a choice of radio buttons or scholastic buttons described, for example, as “aggressive”, “normal” and “mild” respectively. A value of about 0.8 may be taken as a default value.

### Step 2 – Detection

The hexon, both in orientation 1 and orientation 2, is scanned pixel by pixel across the region of interest of the image along its pixel rows and columns. At each position the local radial angular transform is computed and values of  $|c_{j3}|$  and  $\delta_{j3}$  are derived for the two orientations  $j$ . The values of  $\delta_{13}$  and  $\delta_{23}$  under the two hexon orientations are compared to the  $\delta_3$  threshold specified in step 1. If a value does not equal or exceed this threshold the corresponding value of  $|c_3|$  is set to zero. The maximum of the two values  $|c_{13}|$  and  $|c_{23}|$  is then assigned to the test pixel. If only dark or only light scratches have been selected for removal the ratio  $|\text{Imaginary}(c_3) / \text{Real}(c_3)|$  and the quantity  $\text{Real}(c_3)$  are tested against thresholds  $T_1$  of 0.3 and  $T_2$  of 0 respectively, as previously described, to see if the lightness criterion is satisfied. If it is not, the pixel is removed from further

consideration, for example by assigning it a  $|c_3|$  value of zero. At this stage all pixels have  $|c_3|$  values ranging from zero to  $|c_3|_{\max}$ , the largest response anywhere in the search area.

### Step 3 – Refinement

- 5 The operator adjusts two contrast limits  $L_1$  and  $L_2$  to further restrict what region of the image is selected as a defect area. Only values of  $|c_3|$  satisfying the relationship  $L_1 \leq |c_3| \leq L_2$ , where  $0 < L_1 < L_2 \leq |c_3|_{\max}$ , are considered to represent a defect. To assist the operator it is desirable to mark the selected area of the image. This can be done, for instance, by blinking the pixels or by surrounding them with a “marching ants” selection
- 10 marquee. However, a preferred way is simply to display the corrected pixels. The operator can then easily judge if all of the defects have been removed and whether the rest of the image remains undamaged and so arrive at an effective setting of  $L_1$  and  $L_2$ .

### Step 4 – Correction

- 15 Correction of the pixels defining the defect is accomplished by placing a window 6 pixels by 6 pixels over each pixel to be corrected. Using a numbering scheme for the pixels of the window running from 1 to 6 and starting at the top left of the window, the window pixel with coordinates (3,3) is placed over the pixel to be corrected. If the image contains more than one channel, each is corrected in the same way. The mean channel value in the
- 20 window,  $C_{\text{mean}}$ , is computed. A replacement channel value,  $C_{\text{repl}}$ , is calculated for a light defect as the median of those channel values in the window that are less than  $C_{\text{mean}}$ , and for a dark defect as the median of those channel values in the window that are greater or equal to  $C_{\text{mean}}$ .  $C_{\text{repl}}$  is assigned to any pixel in the window whose channel value exceeds  $C_{\text{mean}}$  in the case of a light defect, and does not exceed  $C_{\text{mean}}$  in the case of a dark defect.

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